

CAP 931 – Sales Agent Prototype Documentation

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Project: AI Sales Assistant – Account Intelligence

Duration: 2 Days

1. Technical Setup

This project implements a prototype AI Sales Assistant that generates account intelligence for sales representatives.

Tech Stack

- Python
- Streamlit (UI)
- FLAN-T5 Large (LLM)
- Requests & BeautifulSoup (web data extraction)
- PyPDF (PDF parsing)

Files

- `app.py`: Streamlit UI and input handling
- `agents.py`: Sales agent logic and prompt orchestration
- `utils.py`: Web and PDF text extraction
- `llm.py`: FLAN-T5 summarization logic
- `config.py`: API keys (excluded from GitHub using `.gitignore`)

The application collects inputs via Streamlit, extracts public data from URLs and PDFs, summarizes it using an LLM, and outputs a one-page account intelligence report.

2. Time Management

Task	Time
Environment & Streamlit setup	2 hours
UI development	3 hours
Web & PDF extraction	4 hours
LLM prompt design	3 hours
Debugging & integration	3 hours
Documentation	1 hour
Total	16 hours

3. LLM Model Selection & Use

Model Used: FLAN-T5 Large

Justification:

- Open-source and cost-efficient
- Strong instruction-following and summarization capabilities
- Suitable for structured business insights
- Avoids dependency on paid APIs

Prompt constraints ensure the model only generates sales-related insights.

4. Inputs Handling

The Streamlit interface accepts:

- Product Name (required)
- Company URL (required)
- Product Category
- Value Proposition (required)
- Target Customer
- Competitor URLs
- Optional PDF upload

Validation ensures required fields are completed before report generation.

5. Data Integration & Output Relevance

- Public web data is extracted from company and competitor URLs
- PDFs are parsed to enrich product context
- Summaries are generated using LLM prompts
- Output is a single-page account intelligence report including:
 - Company Strategy
 - Competitor Mentions
 - Leadership Insights
 - Product Fit Summary
 - Sources

6. Challenges and Solutions

Challenge: Web scraping restrictions (403 errors)

Solution: Used corporate “About” pages and added browser headers.

Challenge: Inconsistent LLM responses

Solution: Added fallback summaries and minimum-length checks.

Challenge: API key security

Solution: Stored keys in config.py and added it to .gitignore.

7. Experiments

Different prompt formats were tested. Short, role-focused prompts produced the most relevant and concise one-page summaries.

8. System Outputs

The generated one-page account intelligence report is included via:

- Screenshots of the Streamlit output

Sources

- <https://corporate.walmart.com/about>

- <https://corporate.target.com/about>
- <https://www.aboutamazon.com/about-us>

9. Production Deployment Considerations

For production deployment:

- Host Streamlit on cloud infrastructure
- Use environment variables for secrets
- Add caching, logging, and access controls

Submission Checklist

Streamlit app runs

One-pager output generated

Screenshots included

Documentation file completed

API keys not committed

GitHub repo clean